

URBAN DYNAMICS MODELLING USING ANT NEST BUILDING

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ABSTRACT

Urban dynamics deal with spatial organizations where a great complexity of interactions appears. Social and economic aspects interact and environmental objectives are nowadays a major purpose for sustainable urban development. We propose some generic modelling processes able to face with this complexity, in order to simulate the evolution of the city centers. These organizational centers need a multi-criteria description for their evolution, including feed-back phenomena of them over their environment and components. We propose a swarm intelligence algorithm, using social-insect collective behavior. We combine a decentralized approach, based on emergent clustering mixed with spatial constraints or attractions, as an extension of the ant nest building algorithm with multi-center. Typically, this model is currently used by ourself, to model and analyze cultural equipment dynamics in urban area.

INTRODUCTION

Many natural and artificial systems have emergent properties based on spatial development. This spatial development is both the result of some mechanisms from the system behavior and the actor of the system formation by morphogenetic feedback. Natural ecosystems or social organizations in urban dynamics are typically such emergent spatial organizations. The goal of this paper is to study some computable mechanisms and algorithms able to model such spatial self-organization processes, taking into account the complexity of the phenomena.

In the next section, complex system concepts are defined and their applications to urban dynamics understanding are described, then swarm intelligence algorithms are

described as methodologies to implement the complex systems concepts, using distributed computations. The following section proposes some specific swarm intelligence methods based on ant systems in order to model the spatial organizations emergence. Experiments are given, using RePast multi-agent platform mixed with a geographical information system.

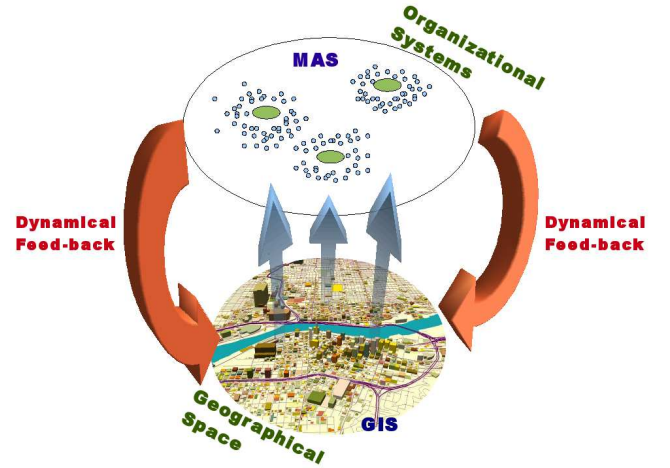


Figure 1: Spatial organizations Complexity Description and the Conceptual Generic Model Based on Swarm Intelligence

MODELLING SPATIAL COMPLEXITY

Complex System Concepts

Complex system theory (7) is based on the fact that for many applicative domains, we can find similar processes linking emergent global behavior and interaction network of constituents. The global behavior is generally not accessible using classical analytical methods.

In classical analytical methods, the global behavior of the system is the description of the equations. Simulations from these formulations, consist in obtaining the trajectories of the behavior predefined in the equation

formulation.

In complex systems modelling, we have to model the constituents of the system and the interaction network or system which link these constituents, using a decentralized approach. So the global behavior of the system cannot be understood by the description of each constituent. In complex system modelling, the global behavior is an emergent property from the interaction network or systems between its constituents and lead to the creation of an dynamical organization of these constituents

This dynamical and emergent organization retro-acts on its own components. Two kinds of feedback allow to describe these phenomenon. The positive feedback means that the emergence increases the organization constitution. the negative feedback means that the emergence has regulator properties which finally stop the increasing organization constitution and allow the system stabilization.

Another major aspect of complex systems is that they can be considered as open systems. This means that they are crossed by energetical fluxes that make them evolve in a continuous way. From these energetical fluxes, complex systems can evolve through critical states, using bifurcation schema and attractors behaviors. One of the major vector or support of these energetical fluxes is the environment itself where the complex systems and their entities evolve. In many natural and artificial systems, the environment has some spatial effects which interact on the whole complexity of the phenomenon. This spatial environment can be modified by the system but he can also be the catalyst of its own evolution. Understanding and modelling the deep structural effect of the interaction between the systems and its spatial environment is the goal of the study presented in the sequel.

Application to urban dynamics

Social and human developments are typical complex systems. Urban development and dynamics are the perfect illustration of systems where spatial emergence, self-organization and structural interaction between the system and its components occur. In figure 1, we concentrate on the emergence of organizational systems from geographical systems. The continuous dynamic development of the organization feed-back on the geographical system which contains the organization components and their environment. The lower part of this figure explains our analysis methodology. It consists to describe many applicatives problems by dynamical graphs or environments in order to detect organizations over these dynamical environment. For the organization detection, we use swarm intelligence

processes. We model the feed-back process of this emergent organization on the system constituents and its environment. To analyse or simulate urban dynamics, nowadays, we can use the great amount of geographical databases directly available for computational treatment within Geographical Information Systems. On the organizational level description, the new development of multiagent systems (MAS) allows nowadays to develop suitable models and efficient simulations.

The applications we focuss on in the models that we will propose in the following concerns specifically the multi-center (or multi-organizational) phenomona inside urban development. As an artificial ecosystems, the city development has to deal with many challenges, specifically for sustainable development, mixing economical, social and environmental aspects. The decentralized methodology proposed in the following allows to deal with multi-criteria problems, leading to propose a decision making assistance, based on simulation analysis.

Cultural dynamics processes in urban can be modelled using such methodology. It is typically a multi-criteria self-organization processes where appears emergent usage oo urban equipment. A modelling of these dynamics is presented latter in this paper.

Swarm Intelligence and Spatial Environment

Decentralized algorithms have been implemented for many years for various purposes. In this algorithm category, multi-agent systems can be considered as generic methods (11). We deal, in this paper, with agents based on reactive architecture and expected to be used inside numerous entity-based systems. The aim of programs using such architectures, is to deal with emergent organizations using specific algorithms called emergent computing algorithms. Swarm Intelligence is the terminology used to point out such reactive agent-based methods where each entity is built with the same basis of behavior, but reacts in autonomous way.

Based on swarm intelligence concepts, Ant systems (3) propose some bio-inspired method from social insects mecanisms. For exemple, Ant Colony Optimization (ACO) methods is a such method family where the basic entities are virtual ants which cooperate to find the solution of graph-based problems, like network routing problems, for example. Using indirect communications, based on pheromon deposits over the environment (here a graph), the virtual ants react in elementary way by a probabilistic choice of path weighted with two coefficients, one comes from the problem heuristic and the other represent the pheromon rate deposit by all

the ants until now. The feed-back process of the whole system over the entities is modelled by the pheromon action on the ants themselves.

Particle Swarm Optimization (PSO) is a metaheuristic method initially proposed by J. Kennedy and R. Eberhart (6) which is also based on swarm intelligence concepts. This method is initialized with a virtual particle set which can move over the space of solutions corresponding to a specific optimization problem. The method can be considered as an extension of a bird flocking model, like the BOIDS simulation from C.W. Reynolds (9). In PSO algorithm, each virtual particle moves according to its current velocity, its best previous position and the best position obtained from the particles of its neighborhood. The feed-back process of the whole system over the entities is modelled by the storage of this two best positions as the result of communications between the system entities.

Other swarm optimization methods have been developed like Artificial Immune Systems (4) which is based on the metaphor of immune system as a collective intelligence process. F. Schweitzer proposes also a generic method based on distributed agents, using approaches of statistical many-particle physics (10).

The method proposed in this paper is based on Ant systems, allowing to deal with self-organization processes emerging from spatial constraints and attractive areas.

MULTI-CENTER ANT NEST BUILDING TO MODEL EMERGENT SPATIAL ORGANIZATIONS

We will describe in this section, the general algorithm which is proposed to model emergent spatial organizations. This algorithm is based on the ant clustering. We introduce pheromon template to spatially control the clustering from local attraction. This method is a decentralized approach which allows to combine multi-center and multi-criteria problem and we will show how we can apply it to model cultural dynamics in urban areas.

Ant clustering

Ant clustering algorithms are inspired by the corpses or larva classification and aggregation that the ants colony are able to do in the real life. The ants are moving inside a closed area and are able to move some material which are randomly put on this area. After a while, and without any kind of centralized coordination, the ants success to create some material clusters.

The algorithm is based on the following and very simple

behavioral rules that each ant implements :

- When an ant is moving without carrying yet material and find some material, the ant will take the material respecting the probability number :

$$P_p = \left(\frac{k_1}{k_1 + f} \right)^2 \quad (1)$$

where f is the material density that the ant perceives locally around itself and k_1 is the threshold. It is easy to check that if $f \ll k_1$ then P_p is near the value 1 and if $f \gg k_1$ then P_p is near the value 0.

- When an ant is moving when carrying some material, the probability to deposit it is computed by :

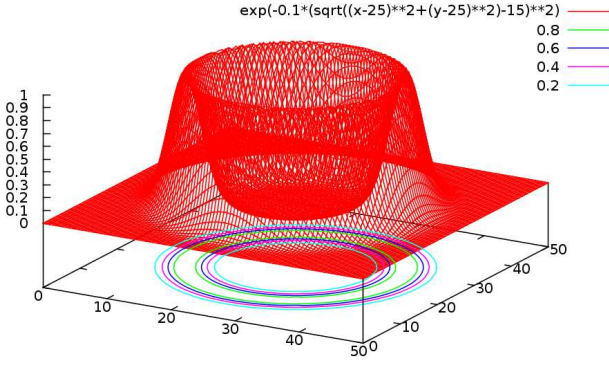
$$P_d = \left(\frac{f}{k_2 + f} \right)^2 \quad (2)$$

where f is still the material density that the ant perceives locally around itself and k_2 is another threshold. It is easy to check that if $f \ll k_2$ then P_d is near the value 0 and if $f \gg k_2$ then P_d is near the value 1.

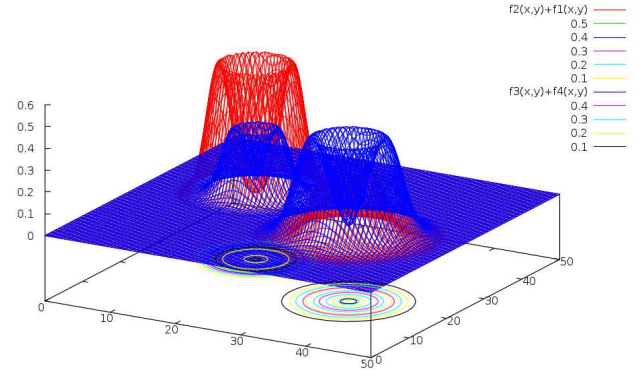
Spatial constraints using template

The ant clustering shows some spatial self-organizations but has the specificity to generate clusters at random places. According to the first random moves that the ants start to do in the beginning of the algorithm, some material will initiate aggregation and the clustering process will complete this aggregation from these initial random first aggregations. To simulate some urban dynamics, we need to introduce specific location with respect to city center for example or cultural equipments. The clustering here will represent the people usage of these centers or equipments and we need to introduce an attractive effect by using a pheromon template. This method follow the algorithm known as Ant Nest Building (3). In ant colonies, the center corresponds to the position of the queen which needs to build the nest and the ant colony moves around it to protect the nest by various material taken on the ground. The queen emits a pheromon which allows to attract the ants during their building. The ant has to deposit the material carried only if the pheromon quantity perceived belongs to a specific range. We use an attractive fonction called P_t , corresponding to a pheromon template and represented by the part (a) of the figure 2.

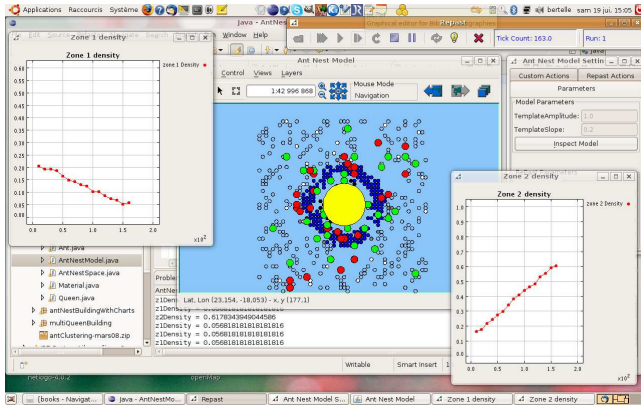
Using this template function, we remplace in the clustering algorithm, the two previous probabilities defined in equation (1) and equation (2) by



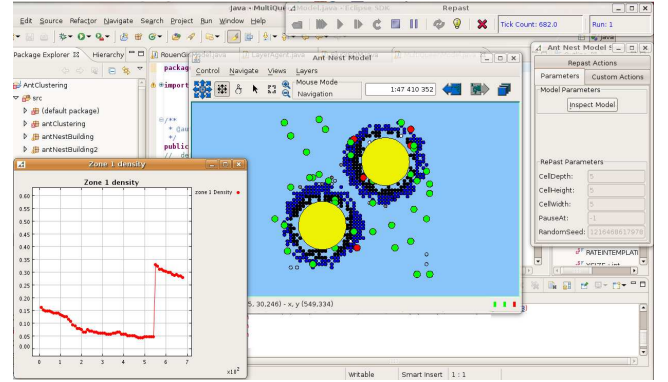
(a) Template function



(a) Template functions



(b) Simulation on RePast



(b) Simulation on RePast

Figure 2: Ant nest building with one center using RePast MAS platform over OpenMap GIS

$$P'_p = P_p(1 - P_t) \quad (3)$$

$$P'_d = P_d P_t \quad (4)$$

In figure 2, we show an implementation of this algorithm using the multi-agent platform called Repast (8). The java version of this platform includes some packages allowing to interface with geographical database and geographical information systems (GIS). The graphical output windows is made under OpenMap which is a GIS developed in Java. In figure, the materials moved by the ants are the small grey circles, the ant moving without material are the green circles, the ant carrying material are the red circles and the queen location is the yellow circle.

Multi-template modelling

The previous subsection describes one local attractive process characterized by the queen and its pheromon template emission. The advantage of this method is to be able to combine the solutions of multi-center and

Figure 3: Multi-center ant nest building using RePast over OpenMap

multi-criteria problems, using interactive processes, each one is represented by a queen and its pheromon template.

On the figure (3), we can see a simulation with two queens and two pheromon templates. It is possible also for each queen to emit many different kinds of pheromons : we called them colored pheromons. Each colored pheromon will attract only the ants associated to its color.

Application to cultural equipment dynamics

The multi-template modelling can be used to model cultural equipment dynamics as described in the figure (4). On this figure, we associate to each cultural center (cinema, theatre, ...) a queen. Each queen will emit many pheromon templates, each template is associated to a specific criterium (according to age, sex, ...). Initially, we put the material in the residential place. Each material has some characteristics, corresponding to the people living in this residential area. The simulation shows the self-organization processus as the result of the set

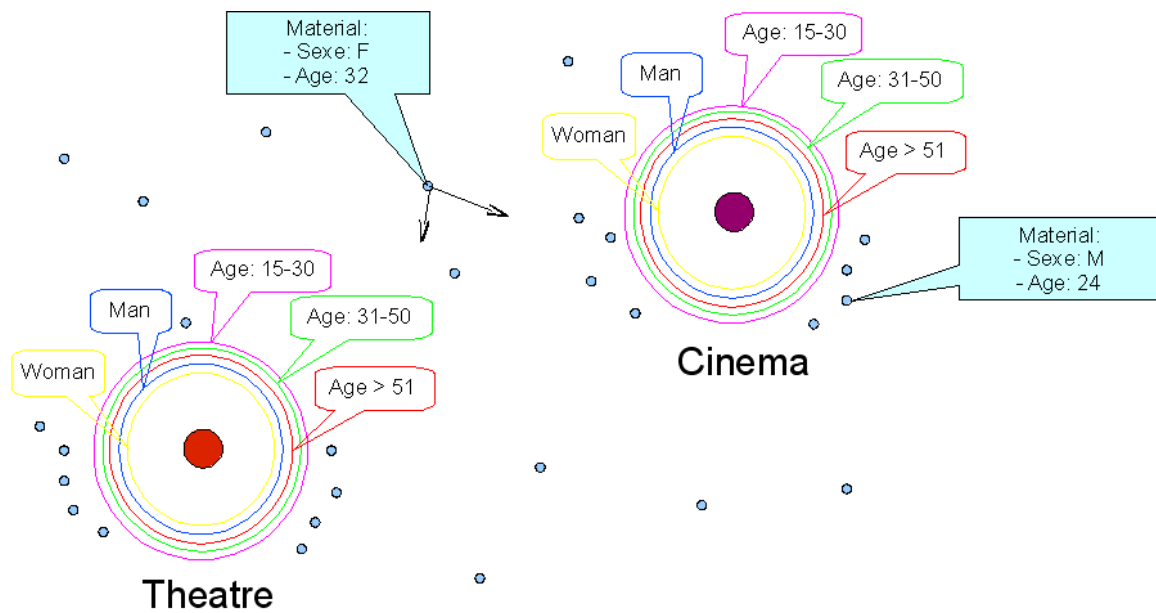


Figure 4: Cultural Equipment Dynamics Modelling

of the attractive effect of all the center and all the templates.

Conclusion and Perspectives

The paper develops some specific swarm intelligence algorithms based on ant colonies processes. Using such decentralized methods, we can model complex multi-center and multi-criteria self-organizations. Urban dynamics are one of the most relevant problems where these approaches can be efficient. These studies are supported by a french regional project (Haute-Normandie) dealing with the study of cultural dynamics over urban area.

REFERENCES

- [1] Benenson, I., Torrens, P.M. (2004) *Geosimulation - Automata-based modeling of urban phenomena*, Wiley.
- [2] C. Bertelle, G.H.E. Duchamp, H. Kadri-Dahmani (eds), *Complex Systems and Self-Organization Modelling*, "Understanding Complex Systems" series, Springer, 2008 (in press).
- [3] Bonabeau, E., Dorigo, M., Theraulaz, G. (1999) *Swarm Intelligence, from natural to artificial systems*, a volume in the Santa Fe Institute Studies in the Sciences of Complexity, Oxford University Press.
- [4] De Castro, L.N., Timmis, J. (2002) *Artificial immune system: a new computational approach*, Springer-Verlag, London, U.K.
- [5] Glnemat, R., Bertelle, C., Duchamp, G.H.E. (2007) *Adaptive Automata Community Detection and Clustering, a generic methodology*, in *Proceedings of World Congress on Engineering 2007, International Conference of Computational Intelligence and Intelligent Systems*, pp 25-30, London, U.K., 2-4th July 2007.
- [6] Kennedy, J., Eberhart, R.C. (1995) *Particle Swarm Optimization* In *Proceedings of IEEE International Conference on Neural Networks (Perth, Australia)*, IEEE Service Center, Piscataway, NJ, 5(3), pp 1942-1948.
- [7] Le Moigne, J.-L. (1999) *La modélisation des systèmes complexes*, Dunod.
- [8] Repast web site (2008) <http://repast.sourceforge.net>.
- [9] Reynolds, C.W. (1987) *Flocks, Herds and Schools: a distributed behavioral model* In *Computer Graphics*, 21(4) (SIGGRAPH'87 Conference Proceedings), pp 25-34.
- [10] Schweitzer, F. (2003) *Brownian Agents and Active Particles*, Springer.
- [11] Weiss, G. (ed.) (1999) *Multiagent Systems*, MIT Press.
- [12] Xiao, N. (2005) *Geographic optimization using evolutionary algorithms* In *8th International Conference on GeoComputation*, University of Michigan, USA.